Using Data Science to Compare Toronto and Manhattan

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- Toronto is the financial capital of Canada and one of the most vibrant and active cities in the world.
- People always compare it with Manhattan, NY and considers it the equivalent for Canada.
- I will use the data collected for both cities as well as the techniques acquired in this course to make a comparison of both cities based in data rather than perceptions.



Data acquisition and cleaning

Data Sources

Toronto

The data for needed to explore, segment, and cluster the neighborhoods in the city of Toronto is not available in the internet in a form ready to be used in data science. I used the data scraped from the Wikipedia page https://en.wikipedia.org/wiki/List of postal codes of Canada: M for this analysis. I used BeautifulSoup to scrape the Wikipedia page mentioned above and get the information related to Toronto's boroughs and neighborhoods.

New York

The information for the city of New York and its 5 boroughs and 306 neighborhoods is readily available via the link https://geo.nyu.edu/catalog/nyu_2451_34572. This link contains a dataset with all the boroughs and neighborhoods of New York as well as their latitude and longitude.

Data acquisition and cleaning

Data Cleaning and Review

Toronto

The data scraped from BeautifulSoup as well as geolocation data from the site http://cocl.us/Geospatial_data/ were combined in a new dataframe called *df_final*.

New York

New York data was available from the site mentioned in the previous slide and came in a form of a JSON file named "newyork_data.json".

0M1BScarboroughRouge, Malvern43.806686-79.191M1CScarboroughHighland Creek, Rouge Hill, Port Union43.784535-79.162M1EScarboroughGuildwood, Morningside, West Hill43.763573-79.18		postalcode		borough
	0	M1B	0	Scarborough
2 M1E Scarborough Guildwood, Morningside, West Hill 43.763573 -79.18	1	M1C	1	Scarborough
	2	M1E	2	Scarborough
3 M1G Scarborough Woburn 43.770992 -79.21	3	M1G	3	Scarborough
4 M1H Scarborough Cedarbrae 43.773136 -79.23	4	M1H	4	Scarborough

In [45]: neighborhoods.head()
Out[45]:

	Borough	Neighborhood	Latitude	Longitude
0	Bronx	Wakefield	40.894705	-73.847201
1	Bronx	Co-op City	40.874294	-73.829939
2	Bronx	Eastchester	40.887556	-73.827806
3	Bronx	Fieldston	40.895437	-73.905643
4	Bronx	Riverdale	40.890834	-73.912585

Using Foursquare to explore the cities

Toronto

Foursquares' API give access to locations and venues near a particular address or city, recommendations on these venues and much more.

Starting with Toronto; I collected the nearby venues for each neighborhood and created a new dataframe called *toronto_venues*.

The results showed a total of 2258 venues and of are represented in the figure below.

In [84]:	<pre>print(toronto_venues.shape) toronto_venues_total=toronto_venues.shape[0] toronto_venues.head()</pre>											
	(2258,	7)										
Out[84]:		Neighborhood	Neighborhood Latitude	Neighborhood Longitude	Venue	Venue Latitude	Venue Longitude	Venue Category				
	0	Rouge, Malvern	43.806686	-79.194353	Wendy's	43.807448	-79.199056	Fast Food Restaurant				
	1	Rouge, Malvern	43.806686	-79.194353	Interprovincial Group	43.805630	-79.200378	Print Shop				
	2 Highla	and Creek, Rouge Hill, Port Union	43.784535	-79.160497	Chris Effects Painting	43.784343	-79.163742	Construction & Landscaping				
	3 Highla	and Creek, Rouge Hill, Port Union	43.784535	-79.160497	Royal Canadian Legion	43.782533	-79.163085	Bar				
	4 Guil	ldwood, Morningside, West Hill	43.763573	-79.188711	Swiss Chalet Rotisserie & Grill	43.767697	-79.189914	Pizza Place				

Using Foursquare to explore the cities

Toronto - continued

Next step was to analyse each Toronto neighborhood. I created a one-hot encoding matrix with 2258 rows (number of venues) and 280 columns (number of unique categories). Next, I grouped rows by neighborhood using the mean of the frequency of occurrence for each category. This resulted in a dataframe with dimensions 100 x 280.

After putting this data in a pandas dataframe and sorting them in descending order, I created a new dataframe with the top 10 common venues in each neighborhood.

8]:	Neighborhood	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue
0	Adelaide, King, Richmond	Coffee Shop	Café	Steakhouse	Bar	American Restaurant	Gym	Cosmetics Shop	Hotel	Burger Joint	Restaurant
1	Agincourt	Chinese Restaurant	Lounge	Sandwich Place	Breakfast Spot	Women's Store	Discount Store	Dog Run	Doner Restaurant	Donut Shop	Drugstore
2	Agincourt North, L'Amoreaux East, Milliken, St	Park	Asian Restaurant	Playground	Women's Store	Donut Shop	Dim Sum Restaurant	Diner	Discount Store	Dog Run	Doner Restaurant
3	Albion Gardens, Beaumond Heights, Humbergate,	Grocery Store	Liquor Store	Sandwich Place	Fried Chicken Joint	Video Store	Coffee Shop	Pharmacy	Pizza Place	Beer Store	Fast Food Restaurant
4	Alderwood, Long Branch	Pizza Place	Coffee Shop	Gym	Skating Rink	Pharmacy	Pub	Dance Studio	Pool	Sandwich Place	Women's Store

Using Foursquare to explore the cities

Manhattan

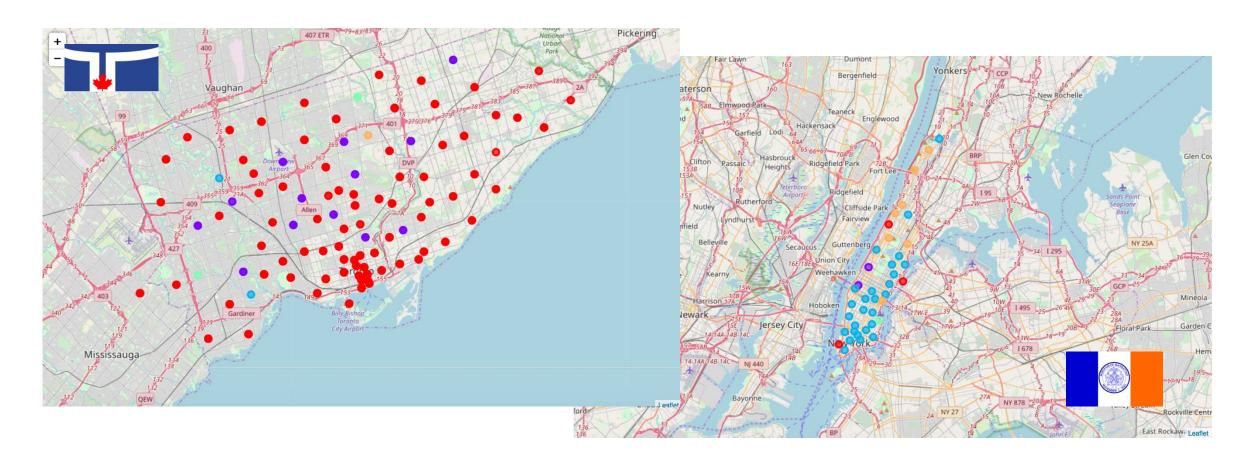
A similar process was followed for Manhattan For Manhattan the dataframe with the top 10 venues in each neighborhood is shown below

10th Mos Commo Venu	9th Most Common Venue	8th Most Common Venue	7th Most Common Venue	6th Most Common Venue	5th Most Common Venue	4th Most Common Venue	3rd Most Common Venue	2nd Most Common Venue	1st Most Common Venue	Neighborhood	
Men's Stor	Plaza	Gym	Clothing Store	Italian Restaurant	Wine Shop	Memorial Site	Hotel	Coffee Shop	Park	Battery Park City	0
Japanes Restaura	Bar	French Restaurant	Cosmetics Shop	Wine Shop	Bookstore	Yoga Studio	Café	Pizza Place	Coffee Shop	Carnegie Hill	1
Liquor Sto	Cosmetics Shop	American Restaurant	French Restaurant	Gym / Fitness Center	Chinese Restaurant	Seafood Restaurant	Art Gallery	Public Art	African Restaurant	Central Harlem	2
Art Galle	Hotel	American Restaurant	Seafood Restaurant	Theater	Nightclub	Bakery	Italian Restaurant	Ice Cream Shop	Coffee Shop	Chelsea	3
Hotpe Restaura	Ice Cream Shop	Vietnamese Restaurant	Dumpling Restaurant	Bubble Tea Shop	Spa	Salon / Barbershop	Cocktail Bar	American Restaurant	Chinese Restaurant	Chinatown	4
Sp	American Restaurant	Sporting Goods Shop	Yoga Studio	Bakery	Sandwich Place	French Restaurant	Coffee Shop	Gym / Fitness Center	Italian Restaurant	Civic Center	5
В	Sandwich Place	Coffee Shop	Spa	Wine Shop	Hotel	American Restaurant	Gym / Fitness Center	Italian Restaurant	Theater	Clinton	6
Steakhous	Taco Place	Gas Station	Café	Convenience Store	Thai Restaurant	Latin American Restaurant	Bakery	Deli / Bodega	Mexican Restaurant	East Harlem	7
Coffee Sho	Ramen Restaurant	Pizza Place	Vegetarian / Vegan Restaurant	Cocktail Bar	Ice Cream Shop	Mexican Restaurant	Chinese Restaurant	Wine Bar	Bar	East Village	8
Ca	Cocktail Bar	American Restaurant	Hotel	Gym / Fitness Center	Bar	Gym	Wine Shop	Steakhouse	Coffee Shop	Financial District	9

Clustering the neighborhoods in both cities

Next step in the analysis is to group the neighborhoods into clusters. I used k-means clustering algorithm to cluster the neighborhoods into 5 clusters (k = 5).

The Folium library was used to visualize the neighborhoods in Toronto and Manhattan and their emerging clusters.



Comparing the two cities - Conclusions

The results of the data from Toronto and Manhattan show several similarities between the two cities:

- An indication that both cities have a rich and vibrant environment is indicated by the number of venues returned for each city: Toronto with 2258 and Manhattan with 3331 venues.
- Both cities are also very diverse and this is demonstrated by the number of unique type venues returned for both; Toronto has 280 and Manhattan has 333 unique type venues.
- On the other hand, considering the difference in size between the cities
 (Manhattan has an area of 33.58 square miles while Toronto is much larger with
 243.33 square miles) it seems like Manhattan has a clear edge over Toronto on
 the number of venues per square miles.
- To summarize; both cities are similar and offer a wide variety of venue choices.
 One can find almost everything they need, or can imagine, in either cities. The
 difference between the two cities is that while in Manhattan one simply need to
 walk around in a neighborhood to find almost all type of venues, in Toronto one
 has to drive or take public transportation since they are spread over a much
 larger surface.

